

Group_inflation_comparison_15-22

February 11, 2025

```
[14]: from inflation_analysis import grouping, calculate_price_indexes
      from tabulate import tabulate
      from tqdm import tqdm
      import matplotlib.pyplot as plt
      import pandas as pd

      # Parameters
      start_year = 2015
      end_year = 2022
      data_folder="/Users/roykisluk/Downloads/Consumer_Expenditure_Survey/"
      top_n = 10
      base_year = start_year
      years=range(start_year, end_year+1)

      groups, total_mmb = grouping(start_year, end_year, cex_data_folder = "
      ↪data_folder)
      groups_mmb = {key: {} for key in groups.keys()}
      for key in groups:
          for year in years:
              groups_mmb[key][year] = groups[key][year][['misparmb']]

      group_analysis = {}
      for key in groups.keys():
          group_number = list(groups.keys()).index(key) + 1
          total_groups = len(groups)
          print(f"Group {group_number}/{total_groups} ({key}) started.")
          combined_df, combined_secondary_df, combined_primary_df, yearly_price_index,
          ↪= calculate_price_indexes(
              start_year, end_year, base_year, group_mmb=groups_mmb[key],
          ↪cex_data_folder=data_folder, verbose=False
          )
          group_analysis[key] = {
              'combined_secondary_df': combined_secondary_df,
              'combined_primary_df': combined_primary_df,
              'yearly_price_index': yearly_price_index
          }
```

```
print(f"Group {group_number}/{total_groups} ({key}) successfully computed.")
```

Group 1/11 (Arabs) started.

Loading price data: 100%| | 8/8 [00:08<00:00, 1.01s/it]

Calculating price indexes: 100%| | 8/8 [00:08<00:00, 1.08s/it]

Group 1/11 (Arabs) successfully computed.

Group 2/11 (Haredi) started.

Loading price data: 100%| | 8/8 [00:06<00:00, 1.17it/s]

Calculating price indexes: 100%| | 8/8 [00:07<00:00, 1.07it/s]

Group 2/11 (Haredi) successfully computed.

Group 3/11 (Low_inc) started.

Loading price data: 100%| | 8/8 [00:06<00:00, 1.18it/s]

Calculating price indexes: 100%| | 8/8 [00:07<00:00, 1.11it/s]

Group 3/11 (Low_inc) successfully computed.

Group 4/11 (High_inc) started.

Loading price data: 100%| | 8/8 [00:06<00:00, 1.17it/s]

Calculating price indexes: 100%| | 8/8 [00:11<00:00, 1.46s/it]

Group 4/11 (High_inc) successfully computed.

Group 5/11 (Young) started.

Loading price data: 100%| | 8/8 [00:06<00:00, 1.18it/s]

Calculating price indexes: 100%| | 8/8 [00:09<00:00, 1.19s/it]

Group 5/11 (Young) successfully computed.

Group 6/11 (Old) started.

Loading price data: 100%| | 8/8 [00:07<00:00, 1.13it/s]

Calculating price indexes: 100%| | 8/8 [00:11<00:00, 1.42s/it]

Group 6/11 (Old) successfully computed.

Group 7/11 (Low_SES) started.

Loading price data: 100%| | 8/8 [00:08<00:00, 1.06s/it]

Calculating price indexes: 100%| | 8/8 [00:07<00:00, 1.09it/s]

Group 7/11 (Low_SES) successfully computed.

Group 8/11 (High_SES) started.

Loading price data: 100%| | 8/8 [00:08<00:00, 1.02s/it]

Calculating price indexes: 100%| | 8/8 [00:04<00:00, 1.73it/s]

Group 8/11 (High_SES) successfully computed.

Group 9/11 (Muslim) started.

Loading price data: 100%| | 8/8 [00:08<00:00, 1.10s/it]

Calculating price indexes: 100%| | 8/8 [00:07<00:00, 1.06it/s]

Group 9/11 (Muslim) successfully computed.

Group 10/11 (Christian) started.

Loading price data: 100%| | 8/8 [00:07<00:00, 1.01it/s]
 Calculating price indexes: 100%| | 8/8 [00:04<00:00, 1.67it/s]
 Group 10/11 (Christian) successfully computed.
 Group 11/11 (Druze) started.
 Loading price data: 100%| | 8/8 [00:08<00:00, 1.03s/it]
 Calculating price indexes: 100%| | 8/8 [00:03<00:00, 2.19it/s]
 Group 11/11 (Druze) successfully computed.

```
[2]: gen_pop_df, gen_pop_secondary_df, gen_pop_primary_df,
      ↪gen_pop_yearly_price_index = calculate_price_indexes(start_year, end_year,
      ↪base_year, cex_data_folder=data_folder, verbose=False)
```

Loading price data: 100%| | 8/8 [00:08<00:00, 1.00s/it]
 Calculating price indexes: 100%| | 8/8 [00:33<00:00, 4.14s/it]

```
[3]: group_counts = {group: {year: len(groups_mmb[group][year]) for year in
      ↪groups_mmb[group]} for group in groups_mmb}
      # Create a dataframe with number of observations per year per group
      observations_df = pd.DataFrame(group_counts).T

      # Calculate the relative share of each group per year
      total_observations_per_year = observations_df.sum(axis=0)
      relative_share_df = observations_df.div(total_observations_per_year, axis=1) *
      ↪100

      # Combine the absolute and relative values into a single dataframe
      combined_df = observations_df.join(relative_share_df, rsuffix='_share')

      # Display the dataframe
      print(tabulate(combined_df, headers='keys', tablefmt='psql'))
```

		2015	2016	2017	2018	2019	2020	2021	
2022	2015_share	2016_share	2017_share	2018_share	2019_share				
	2020_share	2021_share	2022_share						
Arabs	1136	1273	1327	1145	1103	513	951		
727	11.5236	11.9351	12.2825	11.1295	10.9273				
7.49014	11.1476	9.51944							
Haredi	734	778	757	786	565	440	551		
595	7.44573	7.29421	7.00666	7.63997	5.59738				

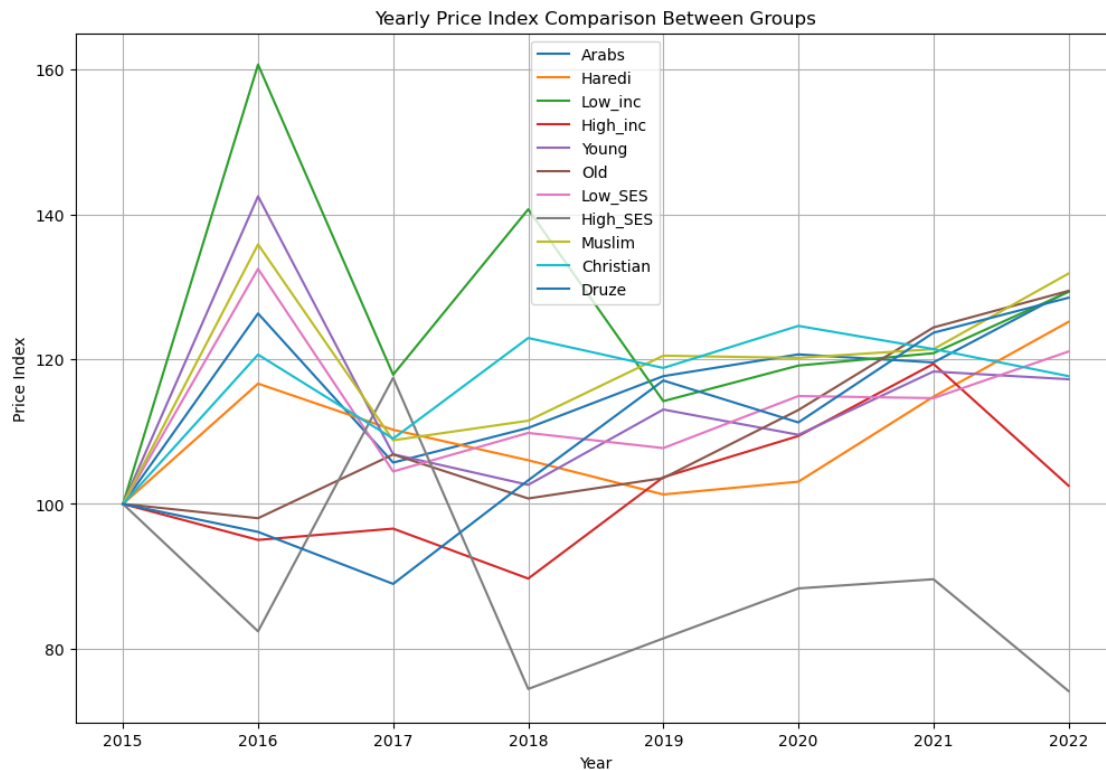
6.4243	6.4588	7.79102						
Low_inc	873	890	940	880	646	397	503	
515	8.85575	8.34427	8.70048	8.55365	6.39984			
5.79647	5.89614	6.74349						
High_inc	1667	1770	1741	1766	1848	1432	1479	
1162	16.9101	16.5948	16.1144	17.1656	18.3079			
20.9082	17.3368	15.2154						
Young	1340	1354	1358	1278	1108	718	877	
820	13.593	12.6945	12.5694	12.4222	10.9768			
10.4833	10.2802	10.7372						
Old	2091	2372	2348	2375	2279	1786	1779	
1663	21.2112	22.2389	21.7327	23.0851	22.5778			
26.0768	20.8534	21.7756						
Low_SES	616	695	728	654	1146	807	1204	
1211	6.24873	6.51603	6.73825	6.35692	11.3533			
11.7827	14.1132	15.857						
High_SES	121	95	102	127	189	163	153	
146	1.22743	0.890681	0.944095	1.23445	1.8724			
2.37991	1.79346	1.91175						
Muslim	880	999	1056	920	870	381	710	
583	8.92676	9.36621	9.77416	8.94246	8.61898			
5.56286	8.32259	7.63389						
Christian	260	309	300	234	231	151	205	
137	2.63745	2.89706	2.77675	2.27449	2.28849			
2.2047	2.403	1.7939						
Druze	140	131	147	123	109	61	119	
78	1.42017	1.2282	1.36061	1.19557	1.07985			
0.890641	1.39491	1.02134						

```
[4]: # Extract yearly price indexes for each group
group_yearly_price_indexes = {group:
    ↪group_analysis[group]['yearly_price_index'] for group in group_analysis}

# Plot the yearly price indexes
plt.figure(figsize=(12, 8))
for group, price_indexes in group_yearly_price_indexes.items():
    years = list(price_indexes.keys())
    indexes = list(price_indexes.values())
    plt.plot(years, indexes, label=group)

plt.xlabel('Year')
plt.ylabel('Price Index')
plt.title('Yearly Price Index Comparison Between Groups')
plt.legend()
```

```
plt.grid(True)
plt.show()
```



```
[17]: # Extract yearly price indexes for each group including general population
group_yearly_price_indexes = {group:
    ↳ group_analysis[group]['yearly_price_index'] for group in group_analysis}
group_yearly_price_indexes['All'] = gen_pop_yearly_price_index

# Define colors for each group
colors = ['#1f77b4', '#ff7f0e', '#2ca02c', '#d62728', '#9467bd', '#8c564b',
    ↳ '#e377c2', '#7f7f7f', '#bcbd22', '#17becf', '#9edae5', '#c7c7c7']

# Create subplots for each year
fig, axes = plt.subplots(nrows=3, ncols=3, figsize=(15, 10), sharey=True)
axes = axes.flatten()

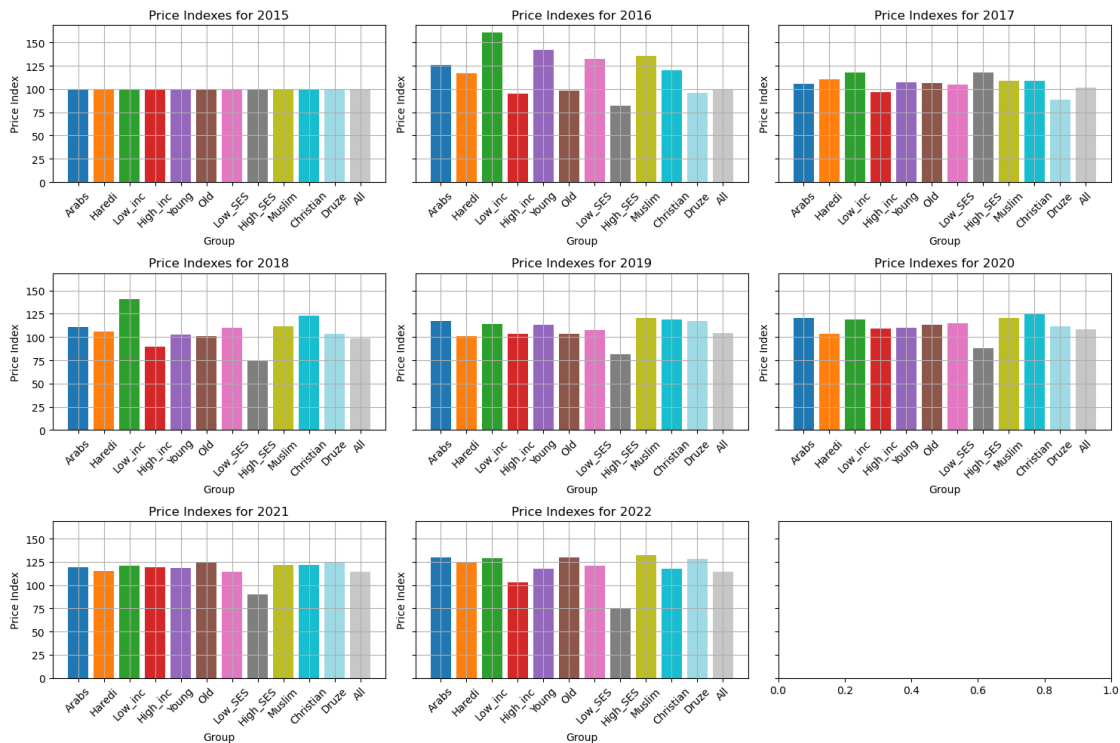
for i, year in enumerate(years[:len(axes)]):
    ax = axes[i]
    groups = list(group_yearly_price_indexes.keys())
    price_indexes = [group_yearly_price_indexes[group][year] for group in
    ↳ groups]
    ax.bar(groups, price_indexes, color=colors)
```

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ax.set_title(f'Price Indexes for {year}')
ax.set_xlabel('Group')
ax.set_ylabel('Price Index')
ax.grid(True)
ax.tick_params(axis='x', rotation=45)

plt.tight_layout()
plt.show()

```



```

[18]: fig, axes = plt.subplots(nrows=4, ncols=3, figsize=(20, 15), sharey=True)
axes = axes.flatten()

for i, (group, data) in enumerate(group_analysis.items()):
    yearly_price_index = data['yearly_price_index']
    years = list(yearly_price_index.keys())
    indexes = list(yearly_price_index.values())

    axes[i].bar(years, indexes, color='skyblue')
    axes[i].set_title(group)
    axes[i].set_xlabel('Year')
    axes[i].set_ylabel('Price Index')
    axes[i].grid(True)

```

```
# Remove any empty subplots
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()
```



```
[19]: for group, analysis in group_analysis.items():
    # Calculate the weight differences between the group and the general
    ↪ population for secondary categories
    weight_diff_df = analysis['combined_secondary_df'].copy()
    weight_diff_df['weight_diff'] = weight_diff_df['weight'] -
    ↪ gen_pop_secondary_df['weight']

    # Sort by the absolute value of the weight differences in descending order
    weight_diff_df['abs_weight_diff'] = weight_diff_df['weight_diff'].abs()
    sorted_weight_diff_df = weight_diff_df.sort_values(by='abs_weight_diff',
    ↪ ascending=False)

    # Display the top_n largest gaps using tabulate
```

```

print(f"Group: {group}")
print(tabulate(sorted_weight_diff_df.head(top_n)[['Year', 'prodcode', 'weight_diff', 'description', 'weight', 'weight_diff']], headers='keys', tablefmt='psql'))
print("\n")

```

Group: Arabs

	Year	prodcode	description	weight	weight_diff
378	2020	395	Personal Jewelry And Watches	0.00219264	-0.135039
314	2019	394	Legal And Other Services	0.00540572	-0.130674
249	2018	392	Personal Services And Cosmetics	0.00436179	-0.129883
119	2016	384	Other Transportation Expenses	0.00644553	-0.126804
371	2020	383	Vehicle Expenses	0.127396	0.11329
184	2017	385	Mail Telephone And Communication	0.0216433	-0.112431
497	2022	383	Vehicle Expenses	0.122325	0.109744
308	2019	383	Vehicle Expenses	0.121403	0.105517
118	2016	383	Vehicle Expenses	0.118808	0.105059
182	2017	383	Vehicle Expenses	0.117387	0.103783

Group: Haredi

	Year	prodcode	description	weight	weight_diff
314	2019	394	Legal And Other Services	0.0045246	-0.131556
508	2022	398	Expenses Not Elsewhere Specified	0.00594857	-0.130851
54	2015	384	Other Transportation Expenses		

0.00228819		-0.130432	
184	2017	391	Cigarettes Tobacco And Smoking Supplies
0.00470577		-0.129368	
249	2018	392	Personal Services And Cosmetics
0.00564004		-0.128604	
119	2016	385	Mail Telephone And Communication
0.00672907		-0.126521	
378	2020	396	Bags Suitcases And Other Products
0.0155506		-0.121681	
443	2021	397	Organization Fees And Donations
0.0130076		-0.0887514	
256	2019	300	Bread Grains and Pastries
0.0745475		0.0693823	
194	2018	302	Meat And Poultry 0.072473
	0.0689943		
+-----+	+-----+	+-----+	+-----+
--+-----+			

Group: High_inc

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	Year	prodcode	description		
weight	weight_diff				
-----+-----+					
---+-----					
505	2022	383	Vehicle Expenses		0.201978
0.19684					
312	2019	383	Vehicle Expenses		0.19702
0.1832					
248	2018	383	Vehicle Expenses		0.195316
0.181458					
378	2020	385	Mail Telephone And Communication		
0.00757866	-0.129653				
249	2018	384	Other Transportation Expenses		
0.00461579	-0.129629				
314	2019	385	Mail Telephone And Communication		
0.00675151	-0.129329				
508	2022	391	Cigarettes Tobacco And Smoking Supplies		
0.00757757	-0.129222				
443	2021	385	Mail Telephone And Communication		
0.00761718	-0.0941419				
376	2020	383	Vehicle Expenses		0.106518
0.0925092					
441	2021	383	Vehicle Expenses		0.105877
0.0917627					
+-----+-----+					
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Group: Young

	Year	prodcode	description		
weight	weight_diff				
508	2022	392	Personal Services And Cosmetics		
0.00873514	-0.128065				
184	2017	384	Other Transportation Expenses		
0.0082545	-0.12582				
311	2019	383	Vehicle Expenses		0.128955
0.123859					
249	2018	385	Mail Telephone And Communication		
0.0168145	-0.11743				

314	2019	391	Cigarettes Tobacco And Smoking Supplies	0.023755
-0.112325				
378	2020	391	Cigarettes Tobacco And Smoking Supplies	
0.0261559	-0.111075			
375	2020	383	Vehicle Expenses	0.115443
0.110277				
247	2018	383	Vehicle Expenses	0.108966
0.0953299				
183	2017	383	Vehicle Expenses	0.108081
0.0942552				
439	2021	383	Vehicle Expenses	0.113704
0.0929				
+-----+				
+-----+				

Group: Old

	Year	prodcode	description	
weight	weight_diff			
119	2016	384	Other Transportation Expenses	
0.0035588	-0.129691			
378	2020	391	Cigarettes Tobacco And Smoking Supplies	
0.0109076	-0.126324			
508	2022	392	Personal Services And Cosmetics	
0.0110679	-0.125732			
314	2019	385	Mail Telephone And Communication	
0.0107218	-0.125358			
249	2018	385	Mail Telephone And Communication	
0.0105258	-0.123719			
184	2017	385	Mail Telephone And Communication	
0.0109945	-0.12308			
312	2019	383	Vehicle Expenses	0.13332
0.1195				
118	2016	383	Vehicle Expenses	0.130615
0.116866				
247	2018	383	Vehicle Expenses	0.130022
0.116385				
504	2022	383	Vehicle Expenses	0.134459
0.11442				

Group: Low_SES

	Year	prodcode	description						
weight	weight_diff								
508	2022	397	Organization Fees And Donations						
0.00208536	-0.134715								
314	2019	392	Personal Services And Cosmetics						
0.00498447	-0.131096								
119	2016	384	Other Transportation Expenses						
0.00683727	-0.126413								
378	2020	393	Personal Products And Cosmetics						
0.0198475	-0.117384								
184	2017	385	Mail Telephone And Communication						
0.0198135	-0.114261								
249	2018	391	Cigarettes Tobacco And Smoking Supplies						
0.0289928	-0.105252								
373	2020	383	Vehicle Expenses			0.110699			
	0.103661								
131	2017	302	Meat And Poultry			0.107198			
	0.100326								
246	2018	383	Vehicle Expenses			0.105049			
	0.100024								
443	2021	395	Personal Jewelry And Watches			0.002078			
	-0.099681								

Group: High_SES

	Year	prodcode	description		weight
weight_diff					
166	2017	362	Dental Care		0.210163
0.208672					
349	2020	362	Dental Care		0.211731
0.198522					
469	2022	362	Dental Care		0.198507
0.198149					
290	2019	362	Dental Care		0.195425
0.190394					
408	2021	362	Dental Care		0.203472
0.181465					
104	2016	362	Dental Care		0.193576

0.178436	
227	2018
362	Dental Care
0.171991	
41	2015
362	Dental Care
0.140754	
184	2017
394	Legal And Other Services
-0.133477	
249	2018
398	Expenses Not Elsewhere Specified
-0.13188	
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Group: Muslim

	Year	prodcode	description	weight	weight_diff
314	2019	395	Personal Jewelry And Watches	0.00843613	-0.127644
492	2022	383	Vehicle Expenses	0.120406	0.119781
430	2021	383	Vehicle Expenses	0.117571	0.114532
249	2018	393	Personal Products And Cosmetics	0.0200521	-0.114192
119	2016	385	Mail Telephone And Communication	0.0209635	-0.112286
367	2020	383	Vehicle Expenses	0.122417	0.109624
181	2017	383	Vehicle Expenses	0.113309	0.108097
244	2018	383	Vehicle Expenses	0.113311	0.105903
194	2018	302	Meat And Poultry	0.108024	0.104545
117	2016	383	Vehicle Expenses	0.115908	0.10238

Group: Christian

	Year	prodcode	description	weight
1	1970	1000	1000	1000
2	1971	1000	1000	1000
3	1972	1000	1000	1000
4	1973	1000	1000	1000
5	1974	1000	1000	1000
6	1975	1000	1000	1000
7	1976	1000	1000	1000
8	1977	1000	1000	1000
9	1978	1000	1000	1000
10	1979	1000	1000	1000
11	1980	1000	1000	1000
12	1981	1000	1000	1000
13	1982	1000	1000	1000
14	1983	1000	1000	1000
15	1984	1000	1000	1000
16	1985	1000	1000	1000
17	1986	1000	1000	1000
18	1987	1000	1000	1000
19	1988	1000	1000	1000
20	1989	1000	1000	1000
21	1990	1000	1000	1000
22	1991	1000	1000	1000
23	1992	1000	1000	1000
24	1993	1000	1000	1000
25	1994	1000	1000	1000
26	1995	1000	1000	1000
27	1996	1000	1000	1000
28	1997	1000	1000	1000
29	1998	1000	1000	1000
30	1999	1000	1000	1000
31	2000	1000	1000	1000
32	2001	1000	1000	1000
33	2002	1000	1000	1000
34	2003	1000	1000	1000
35	2004	1000	1000	1000
36	2005	1000	1000	1000
37	2006	1000	1000	1000
38	2007	1000	1000	1000
39	2008	1000	1000	1000
40	2009	1000	1000	1000
41	2010	1000	1000	1000
42	2011	1000	1000	1000
43	2012	1000	1000	1000
44	2013	1000	1000	1000
45	2014	1000	1000	1000
46	2015	1000	1000	1000
47	2016	1000	1000	1000
48	2017	1000	1000	1000
49	2018	1000	1000	1000
50	2019	1000	1000	1000
51	2020	1000	1000	1000
52	2021	1000	1000	1000
53	2022	1000	1000	1000
54	2023	1000	1000	1000
55	2024	1000	1000	1000
56	2025	1000	1000	1000
57	2026	1000	1000	1000
58	2027	1000	1000	1000
59	2028	1000	1000	1000
60	2029	1000	1000	1000
61	2030	1000	1000	1000
62	2031	1000	1000	1000
63	2032	1000	1000	1000
64	2033	1000	1000	1000
65	2034	1000	1000	1000
66	2035	1000	1000	1000
67	2036	1000	1000	1000
68	2037	1000	1000	1000
69	2038	1000	1000	1000
70	2039	1000	1000	1000
71	2040	1000	1000	1000
72	2041	1000	1000	1000
73	2042	1000	1000	1000
74	2043	1000	100	

weight_diff					
249 2018	398	Expenses Not Elsewhere Specified	0.00292803		
-0.131316					
184 2017	394	Legal And Other Services	0.00567489		
-0.128399					
119 2016	392	Personal Services And Cosmetics	0.00716718		
-0.126083					
300 2019	383	Vehicle Expenses	0.127235		
0.125994					
52 2015	383	Vehicle Expenses	0.135027		
0.121572					
478 2022	383	Vehicle Expenses	0.130368		
0.119949					
314 2020	303	Fish	0.0173376		
-0.118743					
54 2015	385	Mail Telephone And Communication	0.0145448		
-0.118176					
178 2017	383	Vehicle Expenses	0.127629		
0.11199					
417 2021	383	Vehicle Expenses	0.130671		
0.109771					

Group: Druze

weight_diff	Year	prodcode	description	weight
405 2021	383	Vehicle Expenses	0.15883	
0.15128				
236 2018	383	Vehicle Expenses	0.140262	
0.137399				
184 2017	397	Organization Fees And Donations	0.000349413	
-0.133725				
119 2016	392	Personal Services And Cosmetics	0.00436819	
-0.128882				
54 2015	384	Other Transportation Expenses	0.00503555	
-0.127685				
249 2019	303	Fish	0.0084111	
-0.125833				
348 2020	383	Vehicle Expenses	0.135245	
0.124897				

462	2022	383	Vehicle Expenses	0.167994	
0.119967					
361	2021	302	Meat And Poultry	0.117745	
0.116868					
322	2020	332	Electricity Gas And Fuel For Home	0.122888	
0.115144					
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```
[20]: fig, axes = plt.subplots(nrows=4, ncols=3, figsize=(20, 15), sharey=False)
axes = axes.flatten()

for i, (group, analysis) in enumerate(group_analysis.items()):
    # Calculate the weight differences between the group and the general
    ↪ population for secondary categories
    weight_diff_df = analysis['combined_secondary_df'].copy()
    weight_diff_df['weight_diff'] = weight_diff_df['weight'] -
    ↪ gen_pop_secondary_df['weight']

    # Sort by the absolute value of the weight differences in descending order
    weight_diff_df['abs_weight_diff'] = weight_diff_df['weight_diff'].abs()
    sorted_weight_diff_df = weight_diff_df.sort_values(by='abs_weight_diff',
    ↪ ascending=False)

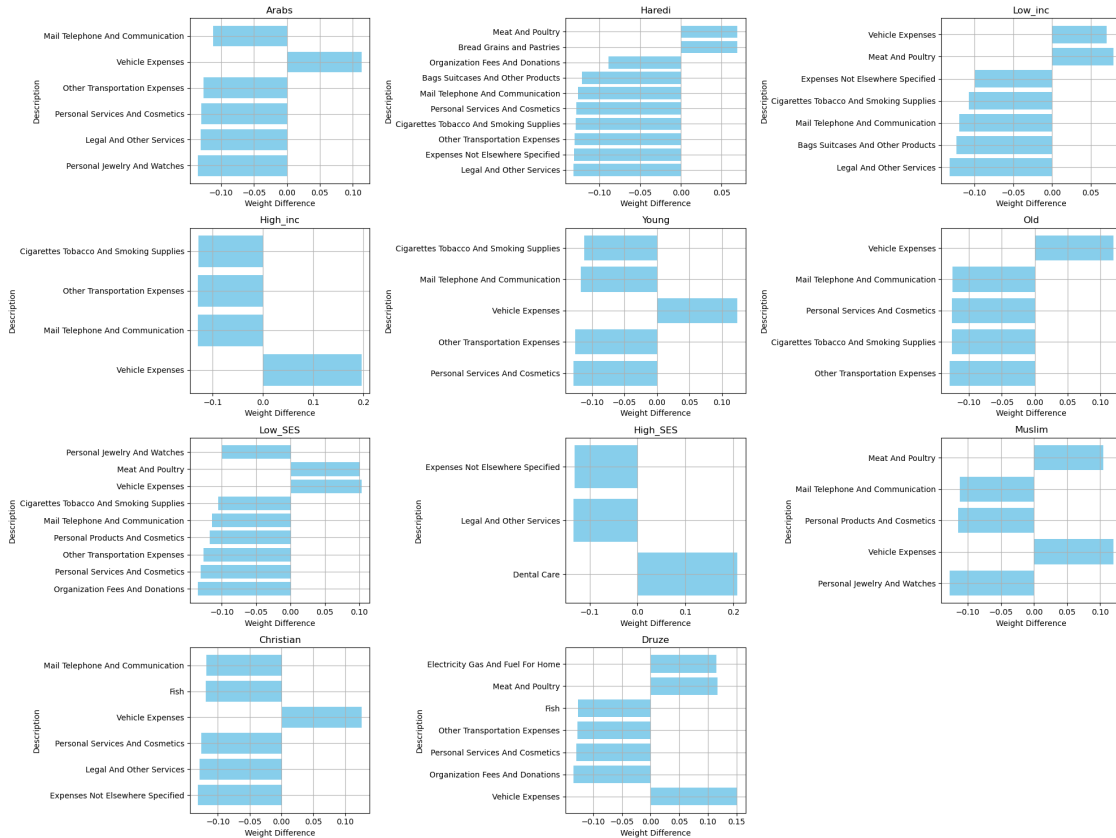
    # Replace NaN values in 'description' with an empty string
    sorted_weight_diff_df['description'] = sorted_weight_diff_df['description'].
    ↪ fillna('')

    # Select the top n largest gaps
    top_n_weight_diff_df = sorted_weight_diff_df.head(top_n)

    # Plot the top n largest gaps
    axes[i].barh(top_n_weight_diff_df['description'],
    ↪ top_n_weight_diff_df['weight_diff'], color='skyblue')
    axes[i].set_title(group)
    axes[i].set_xlabel('Weight Difference')
    axes[i].set_ylabel('Description')
    axes[i].grid(True)

# Remove any empty subplots
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()
```



```
[21]: fig, axes = plt.subplots(nrows=4, ncols=3, figsize=(20, 15), sharey=False)
axes = axes.flatten()

for i, (group, analysis) in enumerate(group_analysis.items()):
    secondary_df = analysis['combined_secondary_df']
    # Filter for increases in price indexes
    increased_price_df = secondary_df[secondary_df['price_index'] > 100]
    # Calculate the contribution to the yearly price index
    increased_price_df = increased_price_df.copy() # Ensure you're working
    ↪with a copy
    increased_price_df.loc[:, 'contribution'] =
    ↪increased_price_df['price_index'] * increased_price_df['weight']
    # Sort by contribution in descending order
    top_contributors = increased_price_df.sort_values(by='contribution',
    ↪ascending=False).head(top_n)

    # Replace NaN values in 'description' with a placeholder text
    top_contributors['description'] = top_contributors['description'].
    ↪fillna('No Description')
```



```

# Plot the top contributors
axes[i].barh(top_contributors['description'],
↳top_contributors['contribution'], color='skyblue')
axes[i].set_title(group)
axes[i].set_xlabel('Contribution to Price Index')
axes[i].set_ylabel('Description')
axes[i].grid(True)

# Remove any empty subplots
for j in range(i + 1, len(axes)):
    fig.delaxes(axes[j])

plt.tight_layout()
plt.show()

```

